CLS Example

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This example shows how to compute Classical Least Squares (CLS) predictions.

CLS Algorithm

The CLS Algorithm is encapsulated in the following MATLAB function.

```
function [C cls, B cls, K cls] = pnnl cls(A train, C train, A unknown)
   %pnnl_cls Classical least squares (CLS) regression
      [C_cls, B_cls, K_cls] = pnnl_cls(A_train, C_train, A_unknown)
   %
   %
       returns concentration matrix C_cls, the least-squares solution that
       minimizes norm(C_cls*K_cls - A_unknown), of the Beer's law
       relationship CK=A. Extinction coefficient matrix K_cls is the
       least-squares solution that minimizes norm(C_train*K - A_train)
   %
   %
       where A train is a matrix of training spectra corresponding to
       known concentrations in the C train matrix. Multiplier matrix
       B_cls is the pseudo-inverse of Beer's law extinction coefficient
   %
       matrix K_cls such that C_cls = A_unknown * B_cls.
   %
   %
       Example:
   %
   %
         load pnnl napalm data
          [C_cls, B_cls, K_cls] = pnnl_cls(A_train, C_train, A_unknown);
   %
   %
       See also pnnl_pcr, pnnl_pls.
   % Copyright 2022-2023 Battelle Memorial Institute
   % Compute K that minimizes norm(CK - A) given C and A
   K_cls = C_train \ A_train;
   % Compute C that minimizes norm(CK - A) given A and K
   C_cls = A_unknown / K_cls;
   % Multiplier matrix B_cls is the pseudo-inverse of Beer's law
   % extinction coefficient matrix K_cls such that
   % C cls = A unknown * B cls.
   B cls = pinv(K cls);
end
```

Concentration Data

The concentrations of the training data are in matrix C_train and the concentrations of the validation data are in matrix C_validation. Column 1 corresponds to the concentrations in constituent 1 (benzene). Column 2 corresponds to the concentrations in constituent 2 (polystyrene). Column 3 corresponds to the concentrations in constituent 3 (gasoline).

```
benzene polystyrene gasoline
                                                           benzene polystyrene gasoline
                          0 100.0000
                                          1
                                                             22.1665 39.0384 38.7951
                                                                                          1
            5.1309
                              94.8691
                                          2
                          0
                                                             21.6874 37.0596 41.2530
                                                                                          2
           10.0660
                          0
                              89.9300
                                          3
                                                             22.1665 39.6980 38.1355
                                                                                          3
           20.1799
                              79.8201
                                          4
                          0
                                                             26.9575 40.3576 32.6849
                                                                                          4
           40.0120
                          0
                              59.9878
                                          5
                                                             25.9993 33.7616 40.2391
                                                                                          5
           59.9972
                          0
                              40.0028
                                          6
                                                             23.1247 37.0596 39.8157
                                                                                          6
                                               C_{\text{validation}} =
                                          7
           79.8412
                          0
                              20.1588
                                                             22.6456 39.0384 38.3160
                                                                                          7
                                          8
           89.8273
                          0
                              10.1727
                                                             22.6456 39.0384 38.3160
                                                                                          8
           100.0000
                          0
                                     0
                                          9
                                                             27.4366 42.3364 30.2270
                                                                                          9
           90.0264
                                    0
                    9.9736
                                         10
C_{\text{train}} =
                                                             27.4366 37.7192 34.8442
                                                                                          10
                                    0
           80.1375 19.8625
                                         11
                                                             26.4784 40.3576 33.1640
                                                                                         11
           64.9950 35.0005
                                     0
                                         12
                                                            26.9575 37.7192 35.3233
                                                                                         12
           21.0228 45.9197
                              33.0575
                                         13
           49.9507
                    5.0599
                              44.9895
                                         14
           40.0182 20.0385
                              39.9433
                                         15
           40.0154 10.0036
                              49.9810
                                         16
           30.0059 10.0282
                              59.9659
                                         17
           40.0340 39.9670
                               19.9990
                                         18
           49.9393
                              46.6859
                                         19
                     3.3748
           46.6501 13.4658
                              39.8840
                                         20
```

Clear variables and load the PNNL napalm data.

```
clearvars
load pnnl_napalm_data
```

Compute CLS

Set up the plot title and color.

```
title_string = sprintf('CLS');
nConstituents = size(C_validation,2);
colorOrder = pnnl_colorOrder(nConstituents);
```

Use all the columns of C_train to compute CLS. Use all the columns of C_validation to compute RMSEP (root mean square error predicted). Use the columns of C_train to compute RMSEC (root mean square error calibration) and RMSECV (root mean square error cross validation).

Compute CLS

```
C_predicted = pnnl_cls(A_train,C_train,A_unknown);
C_calibration = pnnl_cls(A_train,C_train,A_train);
C_cross_validation = pnnl_cross_validation(@pnnl_cls,A_train,C_train);
```

Compute RMSE

```
RMSEP = pnnl_rmse(C_validation,C_predicted);
RMSEC = pnnl_rmse(C_train,C_calibration);
RMSECV = pnnl_rmse(C_train,C_cross_validation);
```

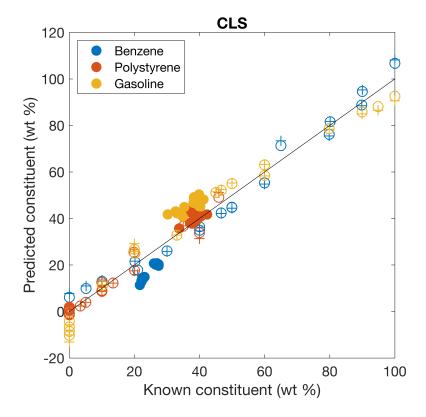
Display RMSE

```
pnnl_display_rmse(title_string,ConstituentNames,...
RMSEC,RMSECV,RMSEP);
```

CLS	Benzene	Polystyrene	Gasoline
RMSEC	4.2721	2.3784	5.6857
RMSECV	5.1043	3.0807	6.7847
RMSEP	7.895	2.6988	9.1105

Plot the results

```
figure
h = gobjects(nConstituents,1);
for k = 1:nConstituents
    % Plot Concentrations
    hold on
    % Validation vs. Predicted
    h(k) = plot(C_validation(:,k),C_predicted(:,k),'.','MarkerSize',35,'Color',colorOr
    % Train vs. Calibration
    plot(C_train(:,k),C_calibration(:,k),'o','MarkerSize',10,'LineWidth',1,'Color',col
    % Train vs. Crosss Validation
    plot(C_train(:,k),C_cross_validation(:,k),'+','MarkerSize',10,'LineWidth',1,'Color
    % 1-1 line
    line(C_train(:,k),C_train(:,k),'Color','k')
    title(title_string)
    xlabel(['Known constituent (',ConcentrationUnits,')'])
    ylabel(['Predicted constituent (',ConcentrationUnits,')'])
    set(gca,'FontSize',14)
    box on
    axis square
    hold off
end
legend(h,'Location','northwest')
```



Legend: Dot is predicted. Circle is calibration. Cross is cross-validation.')

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